

EE698G Assignment 2

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1 Q1

Output Images in Figure 1

2 Q2

See Figure 2 for data

Principal Components obtained to corresponding randomly generated data were : $[0.1912 \ 0.9816]^T$ and $[-0.9816 \ 0.1912]^T$ which is very close to the eigen vectors of covariance matrix $[1 \ 5]^T$, $[-5 \ 1]^T$ as the eigen vectors and principal components are roughly in the same direction.

3 Q3

Answers to theory questions in handwritten form appended after Figure 5.

Output Image on Figure 3 (1.b.1) and Figure 4 (1.b.2 and 1.b.3)

4 Q4

Output Images on Figure 5

5 Q5

See attached handwritten document in the end.

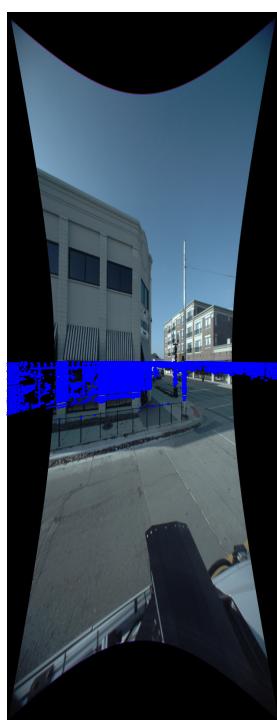
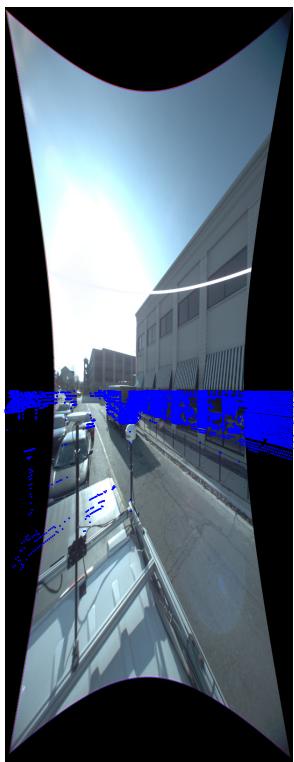
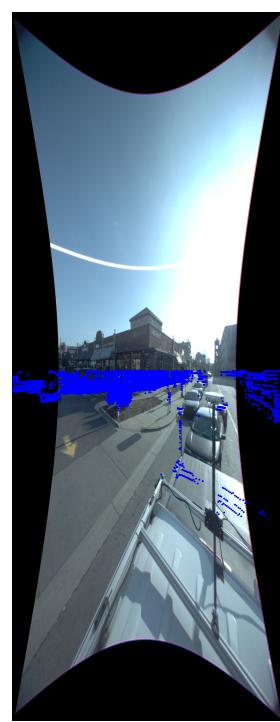
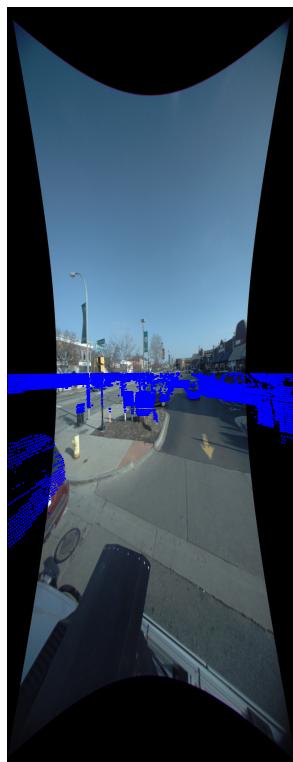
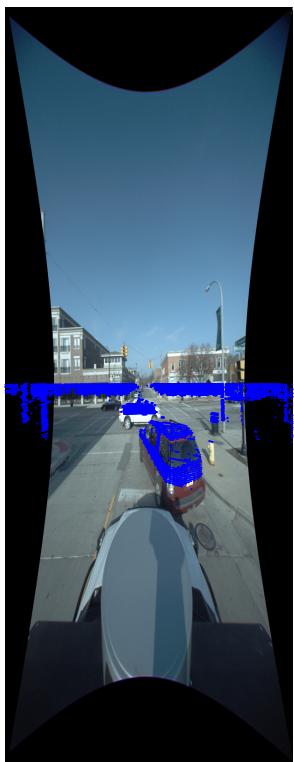


Figure 1: Top Row: 1-3 (Left to Right) Bottom Row 4-5 (Left to Right)

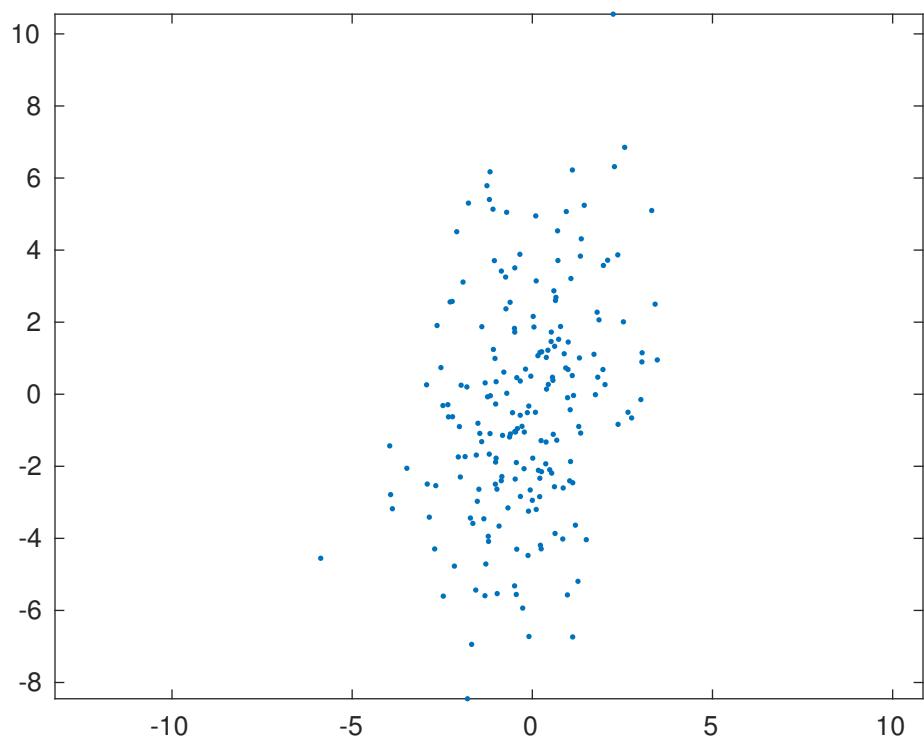


Figure 2: 200 randomly generated points

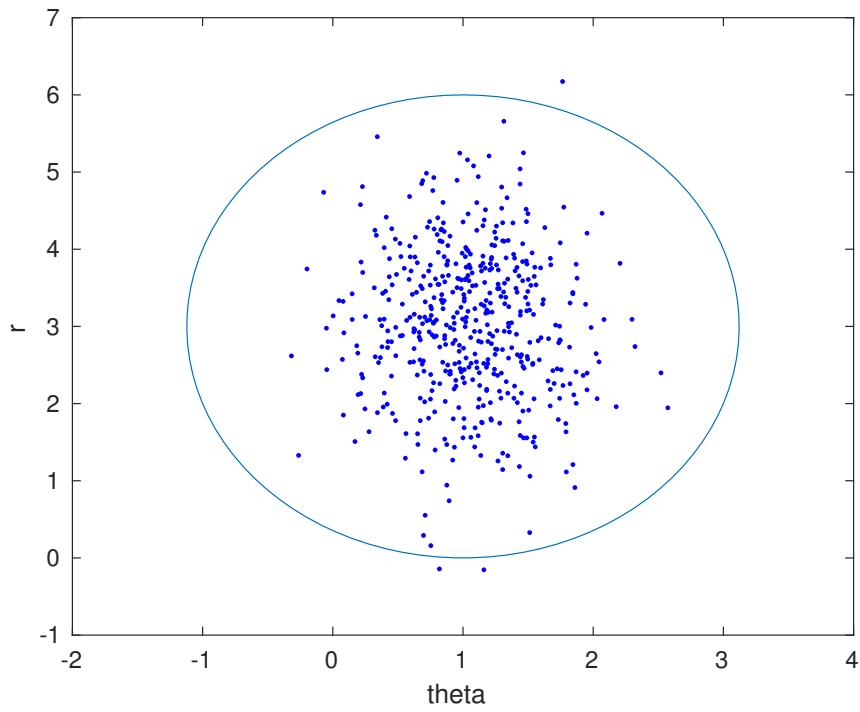


Figure 3: Error ellipse for points in θ -r axis

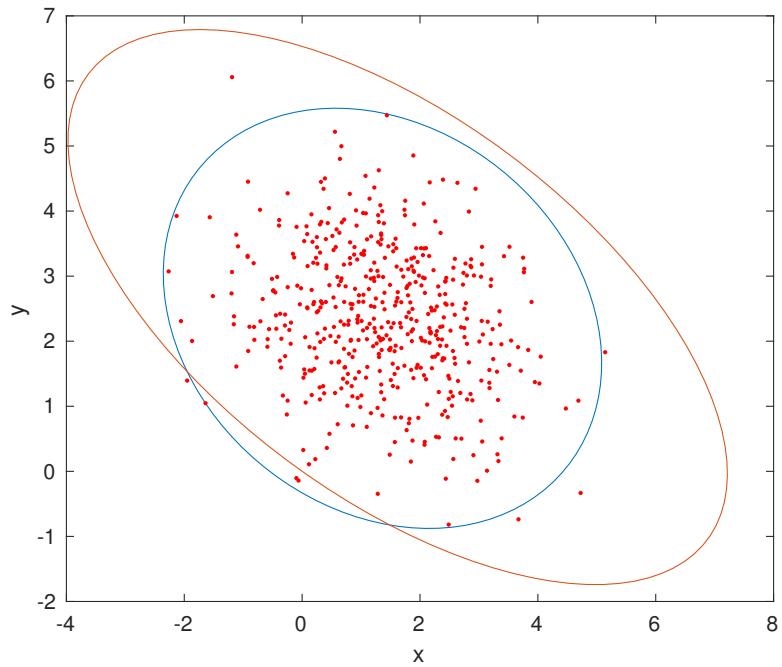


Figure 4: Blue-error ellipse with sample data calculated mean and variance. Red-error ellipse calculated with jacobian (linearisation).

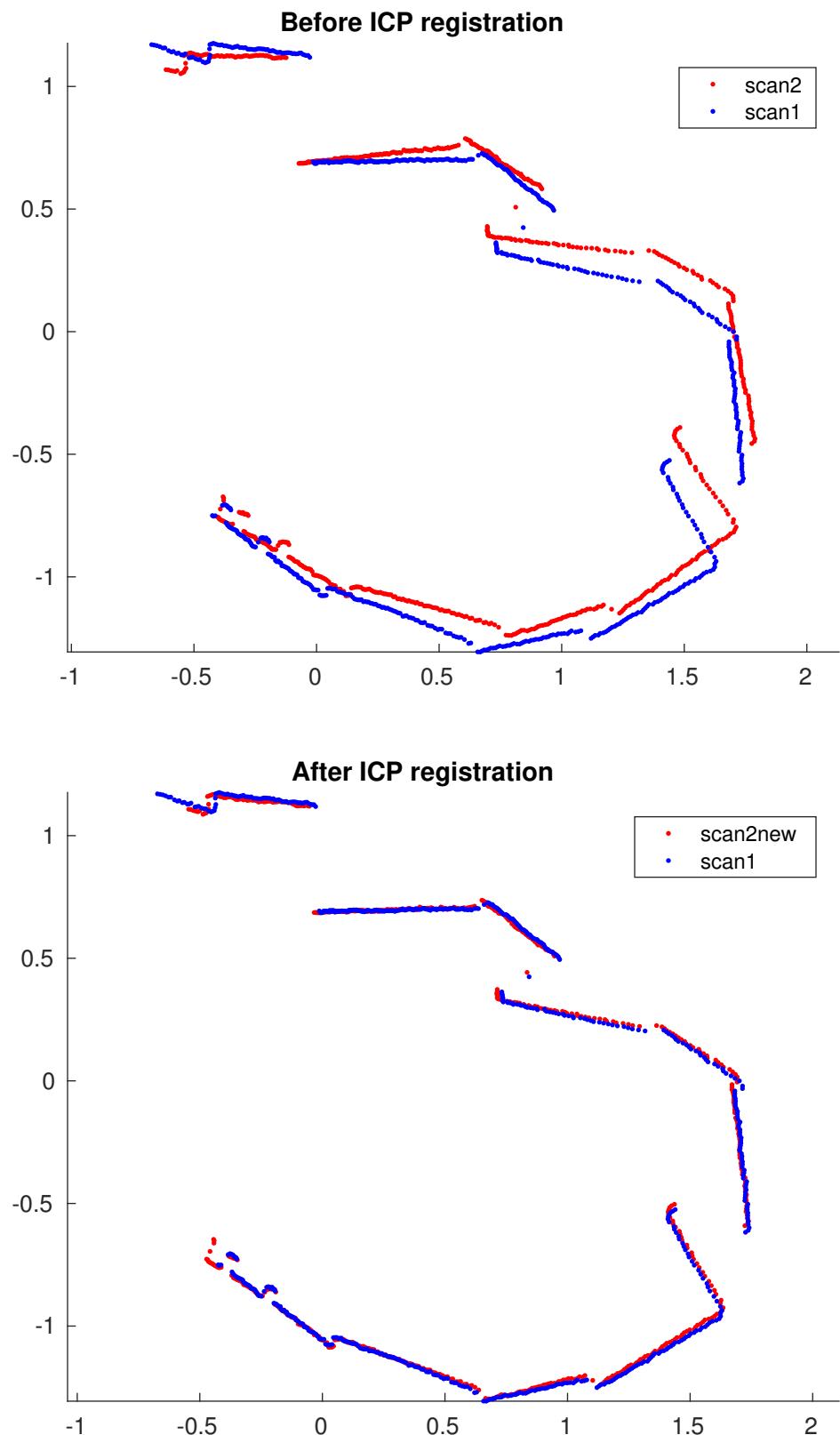


Figure 5: ICP output

Q 3

1.a.1 No, euclidean coordinates need not be normally distributed also, since they are not strictly linear functions of θ and r .

1.a.2

$$J = \begin{bmatrix} \frac{\partial x}{\partial \theta} & \frac{\partial x}{\partial r} \\ \frac{\partial y}{\partial \theta} & \frac{\partial y}{\partial r} \end{bmatrix} = \begin{bmatrix} -r \sin \theta & r \cos \theta \\ r \cos \theta & r \sin \theta \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \end{bmatrix} = J \cdot \left(\begin{bmatrix} \theta \\ r \end{bmatrix} - \begin{bmatrix} \hat{\theta} \\ \hat{r} \end{bmatrix} \right) + \left(\begin{bmatrix} \hat{r} \cos \hat{\theta} \\ \hat{r} \sin \hat{\theta} \end{bmatrix} \right)$$

$\therefore (\theta, r) \rightarrow (x, y)$ is highly non-linear

we choose $\begin{bmatrix} \text{Me} \\ \text{My} \end{bmatrix} = \begin{bmatrix} \hat{r} \cos \hat{\theta} \\ \hat{r} \sin \hat{\theta} \end{bmatrix} = \text{Mc}$

$\begin{cases} \hat{r} = \mu_r \\ \hat{\theta} = \mu_\theta \end{cases}$

$$\begin{aligned} \Sigma_e &= J \Sigma_j J^T \\ &= \begin{bmatrix} -r \sin \theta & \cos \theta \\ r \cos \theta & \sin \theta \end{bmatrix} \begin{bmatrix} 6\theta^2 & 0 \\ 0 & 6r^2 \end{bmatrix} \begin{bmatrix} -r \sin \theta & r \cos \theta \\ r \cos \theta & r \sin \theta \end{bmatrix} \end{aligned}$$

Joint pdf is given by

$$f(x, y) = \frac{1}{\sqrt{(2\pi) \det(\Sigma_e)}} \exp \left(-\frac{1}{2} (\bar{x} - \text{Me})^T \Sigma_e^{-1} (\bar{x} - \text{Me}) \right)$$

where $\bar{x} = [x \ y]^T$ & Me, Σ_e as defined above

1.a.3

Σ_e is defined by linearisation as shown in 1.a.2

$$\begin{aligned}\Sigma_c &= J \Sigma J^T \\ &= \begin{bmatrix} -\sin\theta & \cos\theta \\ \cos\theta & \sin\theta \end{bmatrix} \begin{bmatrix} 6\sigma^2 & 0 \\ 0 & 6r^2 \end{bmatrix} \begin{bmatrix} -\sin\theta & \cos\theta \\ \cos\theta & \sin\theta \end{bmatrix}\end{aligned}$$

1.b.3

This is same as question 1.a.3

1.b.4

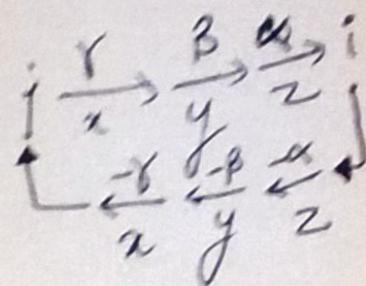
The error ellipse obtained after linearisation is significantly different due to a different covariance matrix. The covariance matrix obtained from linearisation is not accurate because mapping from polar to euclidean coordinates is a highly non-linear on

If plot was made by choosing mean from linearisation the ellipse would have been offset more and more inaccurate

Q5.

To prove: $H_{ij} = \text{inv}(H_{ji})$ Rotation sequence

$$H_{ij} = \begin{bmatrix} R_{ij} & T_{ij} \\ 0 & 1 \end{bmatrix}_{3 \times 3 \quad 3 \times 1}$$

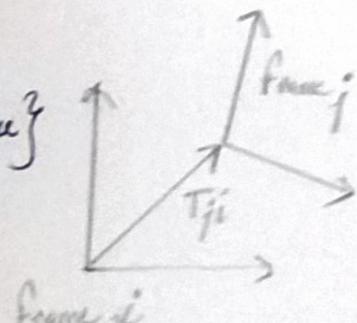


$$H_{ji} = \begin{bmatrix} R_{ji} & T_{ji} \\ 0 & 1 \end{bmatrix}_{3 \times 3 \quad 3 \times 1}$$

$$R_{ij} = R_{ji}^{-1} = R_{ji}^T \quad [\text{Proof shown below}]$$

$$\begin{aligned} R_{ij} &= R_z(\alpha) R_y(\beta) R_x(\gamma) \Rightarrow R_{ji} = R_x(-\gamma) R_y(-\beta) R_z(-\alpha) \\ \Rightarrow R_{ji}^T &= R_z(-\alpha)^T R_y(-\beta)^T R_x(-\gamma)^T \quad \{\text{Euler rotations in reverse}\} \\ \Rightarrow R_{ji}^T &= [R_z(\alpha)]^T [R_y(\beta)]^T [R_x(\gamma)]^T = R_z(\alpha) R_y(\beta) R_x(\gamma) \\ \Rightarrow R_{ji} &= R_{ij} \end{aligned}$$

$$T_{ji} = -R_{ij}^T T_{ij} \quad \{\text{from figure}\}$$



$$\therefore H_{ji} = \begin{bmatrix} R_{ij}^T & -R_{ij}^T T_{ij} \\ 0 & 1 \end{bmatrix}$$

$$\therefore H_{ij} \cdot H_{ji} = \begin{bmatrix} R_{ij} R_{ji}^T & R_{ij} T_{ji} + T_{ij} \\ 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} I_{3 \times 3} & -R_{ij} R_{ij}^T T_{ij} + T_{ij} \\ 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} I & -T_{ij} + T_{ij} \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} I_{3 \times 3} & 0_{3 \times 1} \\ 0 & 1 \end{bmatrix}$$

$$\Rightarrow H_{ij} H_{ji}^{-1} = I$$

$$\Rightarrow H_{ji} = H_{ij}^{-1}$$